

Lecture 2

CS 6260 Automated Planning and Learning

Deterministic Search

Department of Computer Science and Engineering

Indian Institute of Technology Madras



Intro and Logistics (Recap)

Instructor: Pulkit Verma

- Preferred mode of contact:
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- PhD from **Arizona State University**
- Postdoc from **MIT CSAIL**
- Research Areas:
Automated Planning, Human-AI Collaboration, AI Safety and Assessment, Robotics, Lifelong Learning, ...



pulkitverma.net

Teaching Assistants



Sayak



Swapnil



Tushar

Lectures

- CS 15, CSB Ground Floor (of course!!)
- Tu: 11-11:50 AM; W: 10-10:50 AM; Th: 08-08:50 AM; F: 05-05:50 PM
 - Some classes will be cancelled due to official travel; watch out for announcements.
- Attendance is mandatory
- **Office Hours:** Starting Feb 1st week. Watch out for announcement.

Resources

- Course Website:
hails-group.github.io/courses/cs6260/jm26.html
 - Slides posted here
- Announcements: Moodle
- Course Email: cs6260.jm26@gmail.com
 - No URGENT email responses. Please **plan** ahead.
 - We will respond to each email within 48 hours.
 - Any course-related query should be posted on Moodle so that others can also benefit from the answer.



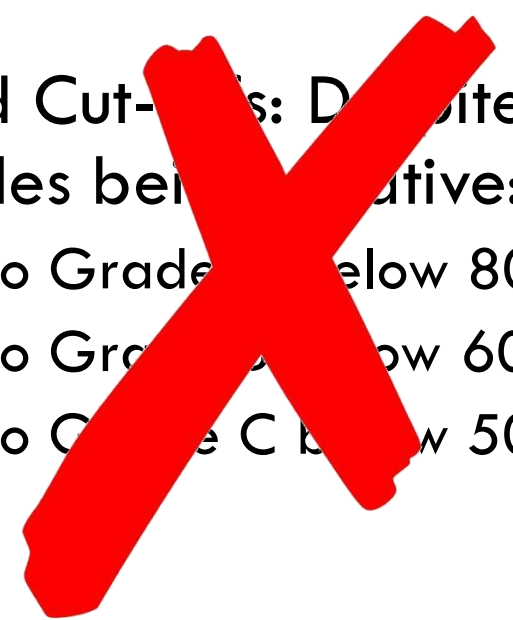
Exams

- 3 Exams:
 1. [Quiz 1] Feb 19: 08:00 AM – 08:50 AM
 2. [Quiz 2] Apr 02: 08:00 AM – 08:50 AM
 3. [End Sem] May 12: 08:00 AM – 08:50 AM
- More details closer to exams.

Grading Breakup

- Exams: 70%
 1. Quiz 1 : 20%
 2. Quiz 2 : 20%
 3. End Sem: 30%
- Assignments: 25%
- In-class Tasks: 5%

- Relative Grading

- Hard Cut-offs: Do not
grades below relative:
 - No Grade below 80%
 - No Grade below 60%
 - No Grade C below 50%
- 

Assignments

- n ($n=4$ or 5) Assignments. Equal distribution, i.e., each assignment will be of $(25/n)\%$.
- Life Happens!!
 - Late submission (up to 2 days) allowed on any one. No questions asked. Just let us know!!
 - If facing any emergency, let us know; we can work something out. We are here to help.

Prerequisites

- Probability
- First Order Logic / Logic in CS
- Programming (Python)
- Analysis of Algorithms

- Sample Knowledge Check:
<https://forms.gle/PnMCm4qy2DuLxQaq7>

- Option to drop the course if you feel like its not for you



Academic Integrity

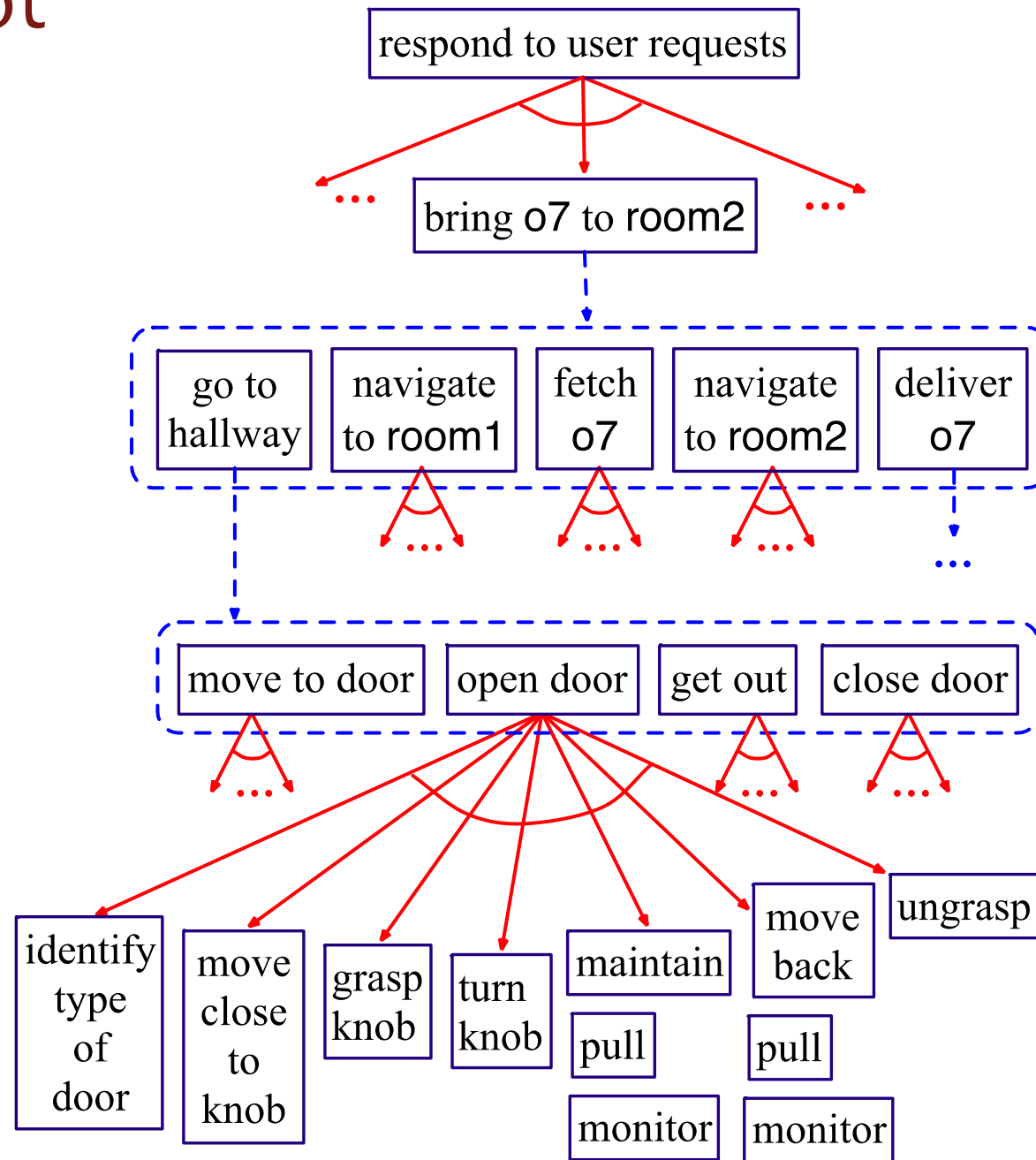
- Ask us if you ever need anything.
 - Office Hours, in addition to regular classes, if you are stuck.
 - Post on Moodle or Email us: No question is too silly or obvious.
 - Collaboration is okay. But final answers should be yours. Cite any collaborators.
 - Do not share your answers with anyone, and do not post any solutions online.
- No relief if any violation found!!
 - Will receive 0 on the assignment/quiz/exam.
 - U grade, if needed.
- Drop the course if you feel you don't know

We are here to Help

- Ask us if you ever need anything.
 - Office Hours, in addition to regular classes, if you are stuck.
 - Post on Moodle or Email us: No question is too silly or obvious.
- Wellness Center: <https://dost.iitm.ac.in/iitmdost/wellness-center>
- Student Wellness Resources: <https://dost.iitm.ac.in/iitmdost/student-wellness-resources-overview>

Recap: Service Robot

- Multiple levels of abstraction
 - Higher levels: more planning
 - Lower levels: more acting
- Continual online planning
 - What room is o7 in?
 - What route?
 - What kind of door?
 - Close enough to door handle?
- Heterogeneous reasoning
 - planning abstract tasks
 - path planning
 - reactive (e.g., open door)



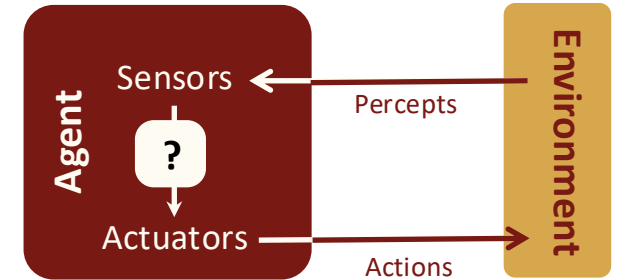
Planning

Acting

Agents

Agents

- **Agent function**
 - Maps percept histories to actions:
 - $f: P^* \rightarrow A$
- Agent function needs to be computed
 - **Agent program:** the program that computes f for a given agent and agent function
- Agent behavior is encoded by the agent function



The Typical Agent Design Process

1. **Represent** background information and objective

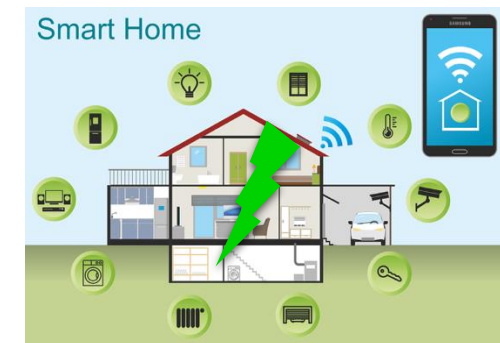
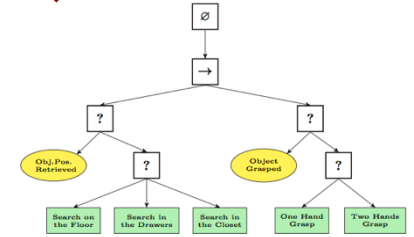
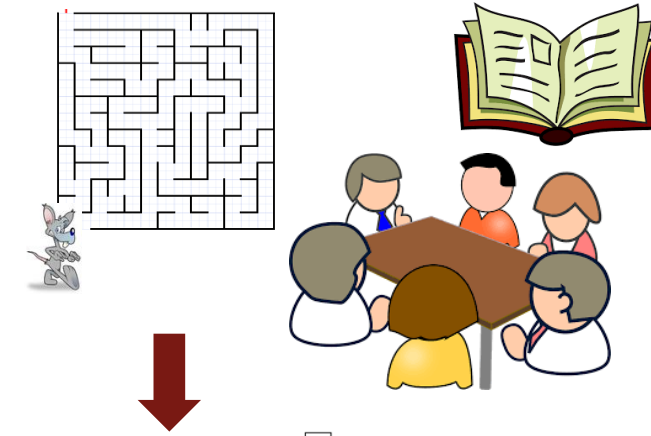
- Multi-step agents require states, actions and their effects on states
- Learned using simulator, modeled using literature/experts, or partially modeled with scope for online learning

2. **Compute** “behavior” that would allow agent to achieve objective

- Behavior may include contingencies depending on observed system state during execution

3. **Execute** this behavior

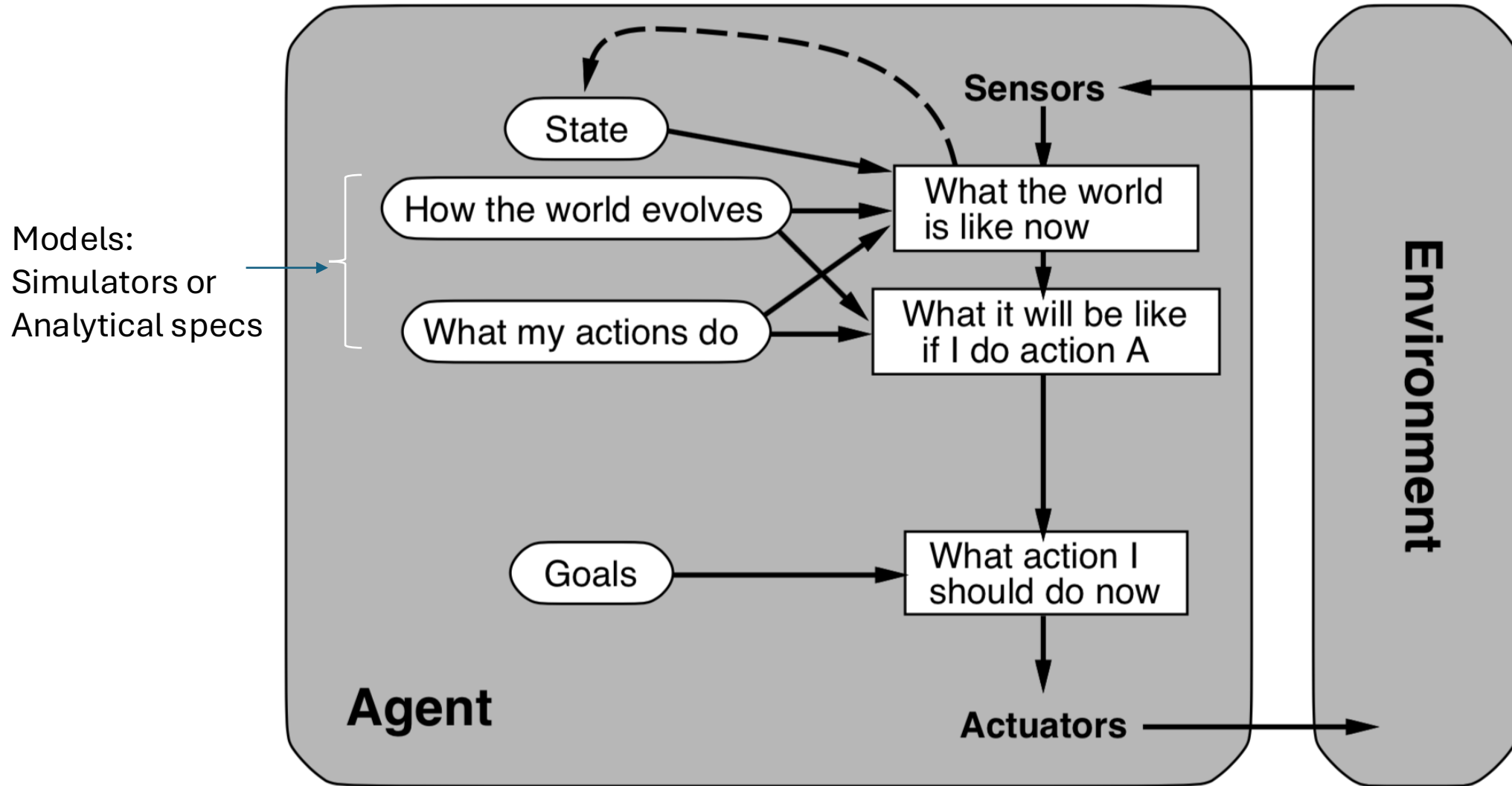
- The bottom line, drives all of the above



Model-based Agents and Model-free Agents

- Model-free agents evaluate courses of action directly through interaction with the real world
 - Rare, due to safety and quality concerns
- Model-based agents evaluate courses of action “in their heads”
 - May employ different types of models:
 - simulators (arbitrary generative models) that provide samples from the possible results of an action
 - analytical models that provide precise sets of possible results of actions
 - E.g., Atari-game-playing agents, AlphaGo, etc.
- Ideal situation: initialize agents using available models; improve through learning on-the-fly

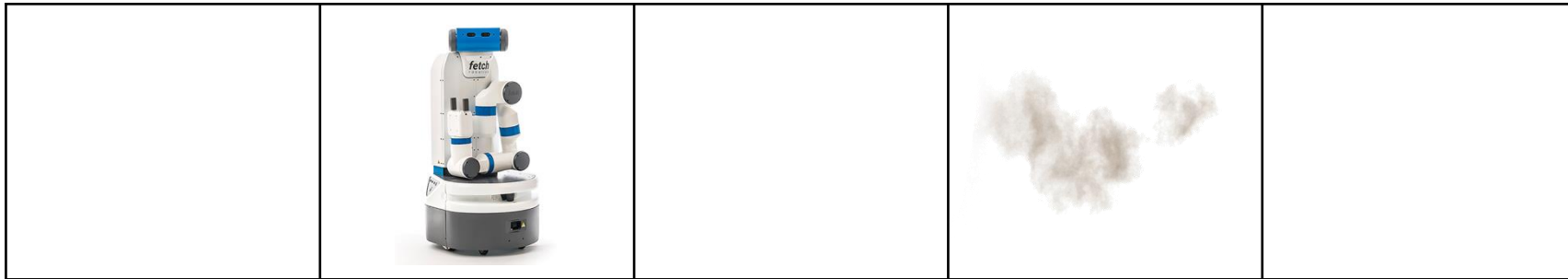
Goal-based Agents



Types of Agents and Environments

Problem Types and Solution Representations

- Deterministic, fully observable problem
 - Actions are deterministic; agent observes full state
 - “Classical planning” or search: solution is a plan, or a sequence of actions



Solution plan: right, right, suck

Forms of Solutions

- Deterministic, fully observable
- Deterministic, non-observable
 - “Conformant planning”: solution is a plan, but often no solutions exist



???



Right, Right, Right, Right, Suck, Left, Suck, Left, Suck, Left, Suck, Left, Suck

Forms of Solutions

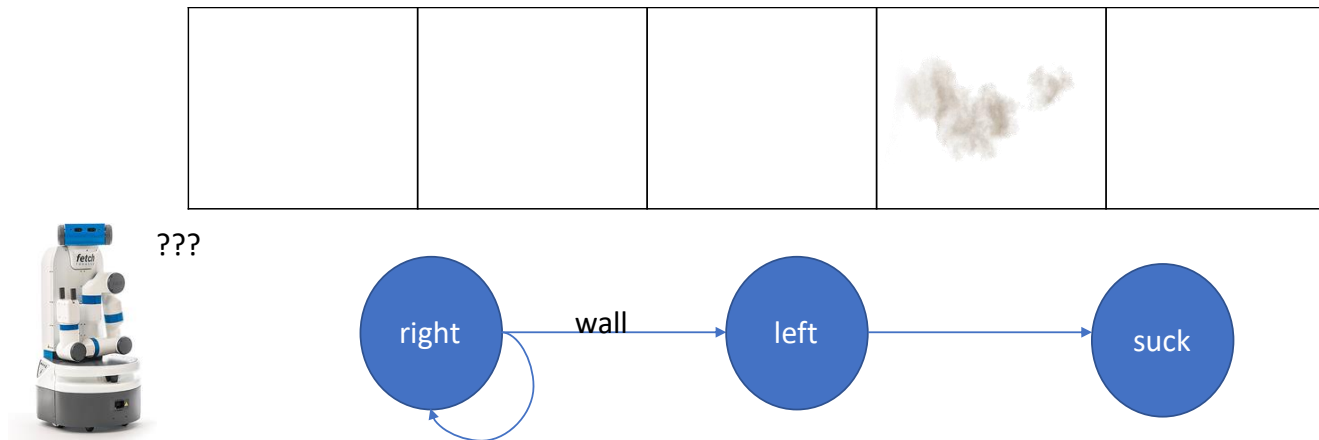
- Deterministic, fully observable
- Deterministic, non-observable
- Non-deterministic, fully observable:
 - Actions may have multiple possible outcomes (noisy actuators, exogenous events)
 - “Contingent planning”: solution can be a policy (a function $S \rightarrow A$)

Non-determinism: a move action may fail



Forms of Solutions

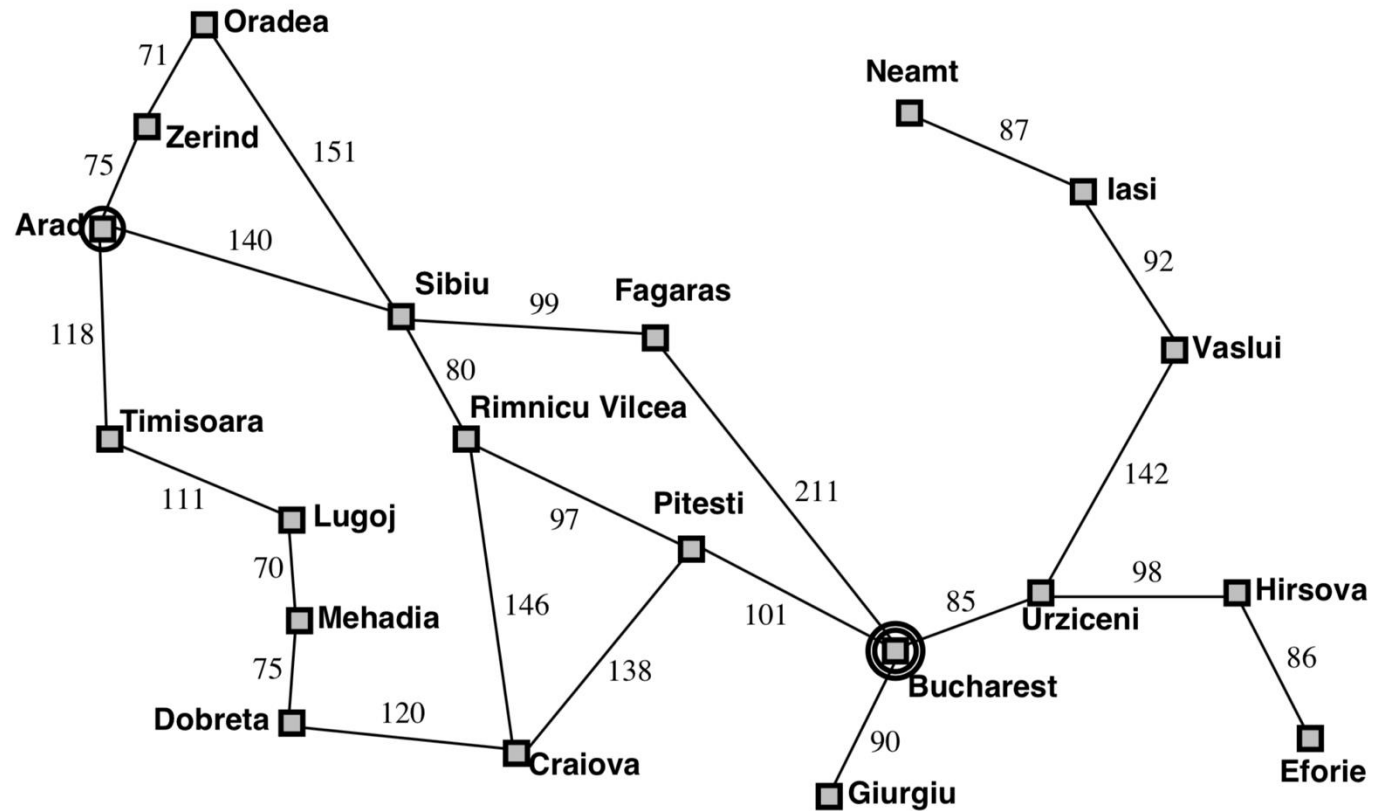
- Deterministic, fully observable
- Deterministic, non-observable
- Non-deterministic, fully observable:
- Partially observable
 - Sensors report noisy, incomplete information about the current state
 - Solution is a finite state machine



Deep Dive: Deterministic and Fully Observable Systems

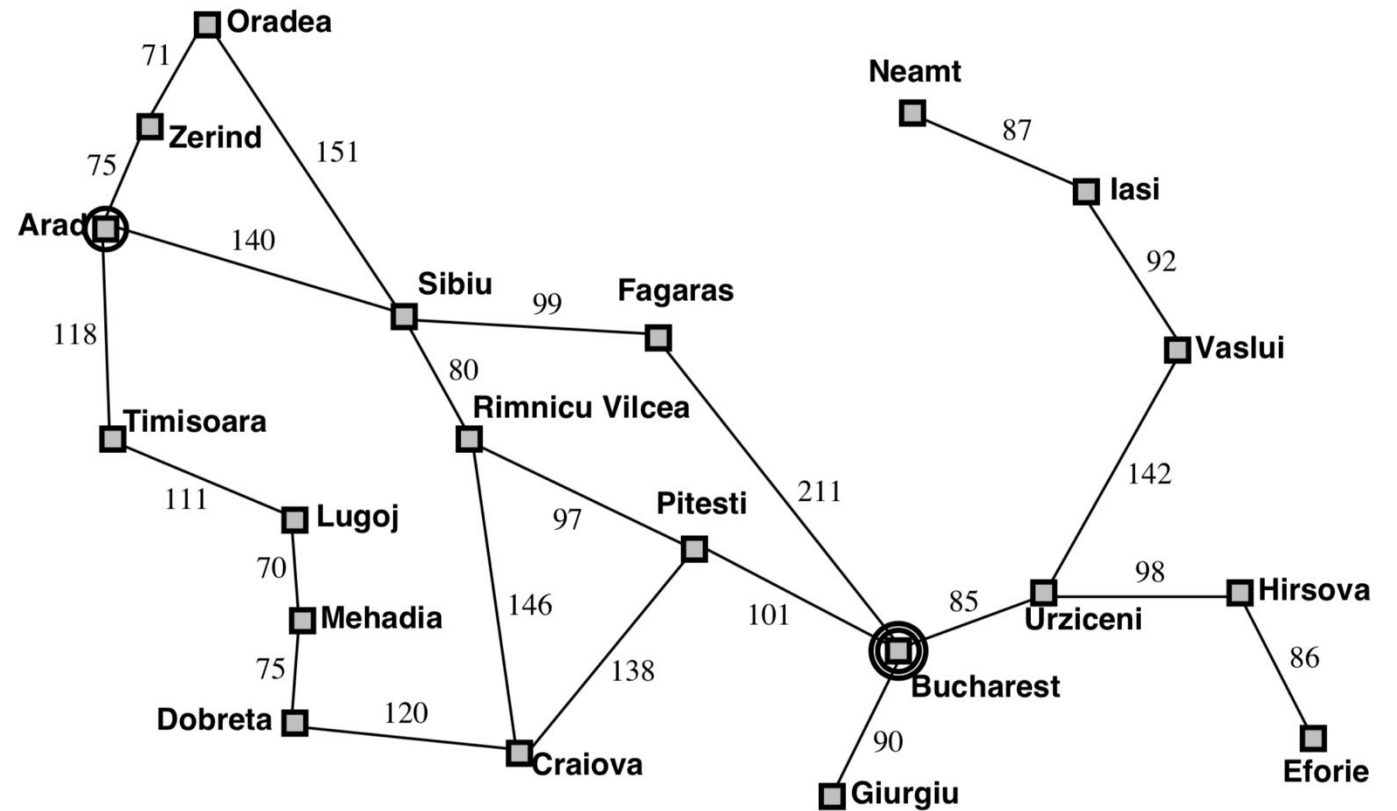
Route-Planning Agent

- State:
 - $\langle \text{current city} \rangle$
- Action:
 - Move along edge
 - Transition function:
 - $T: S \times A \rightarrow S$
- Performance measure:
 - Cost (length) of path from initial city to goal city
- Such problems are known as *search problems*



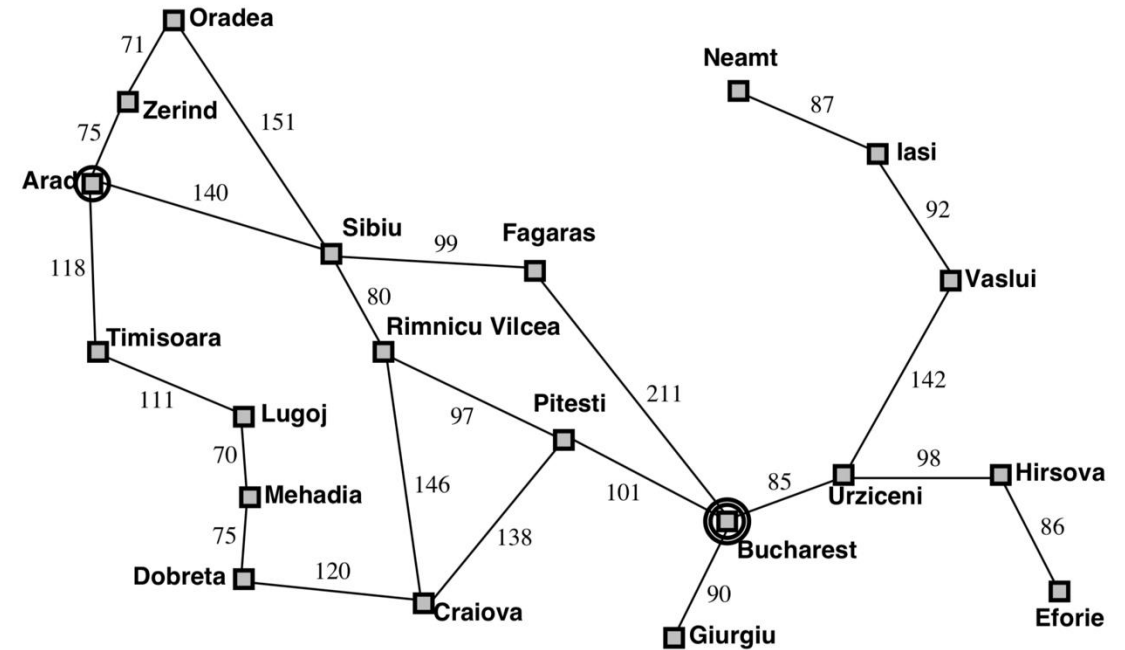
Route-Planning Agent

- Examples of search algorithms?



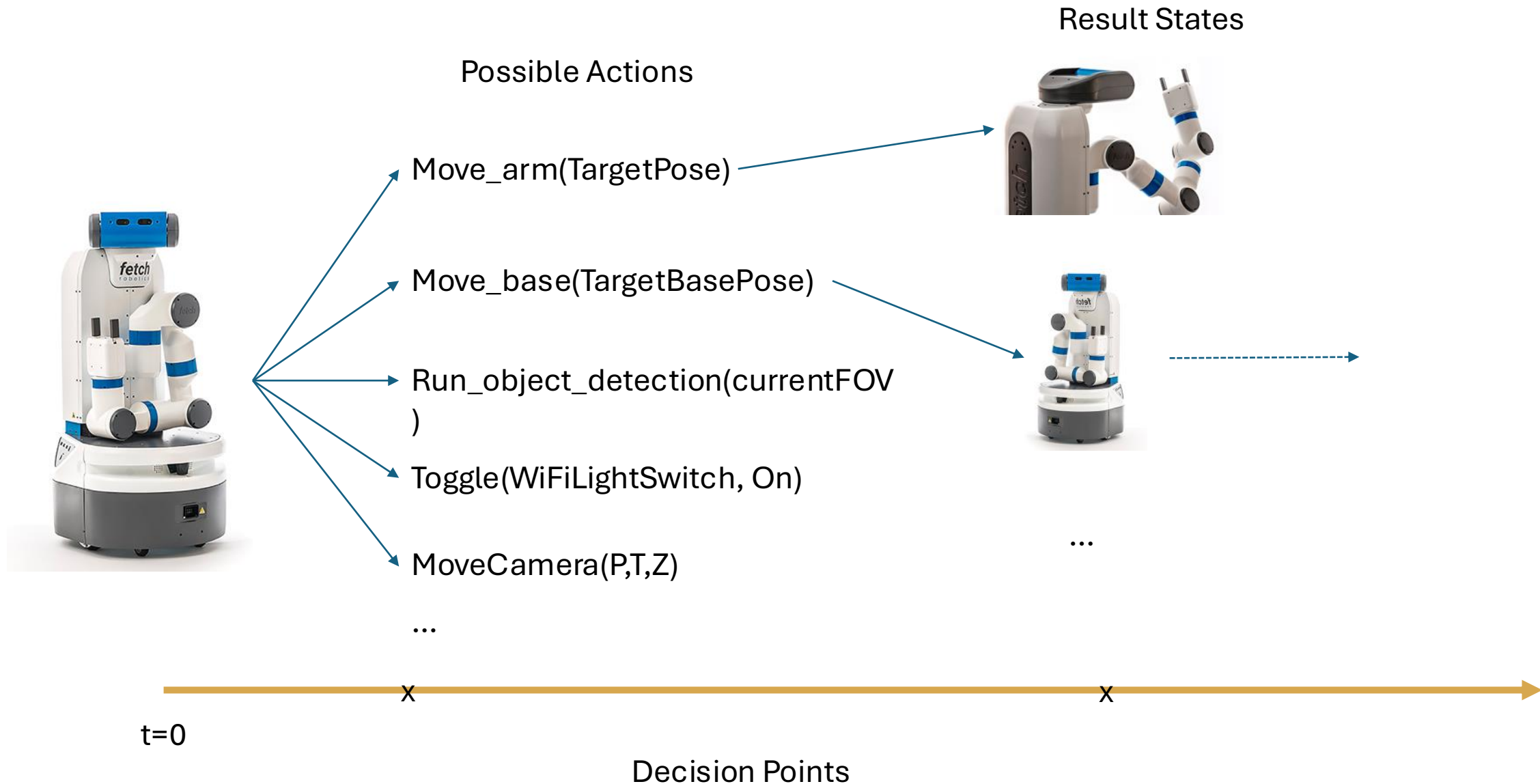
Scope of Search Problems

- Search problems go beyond route planning
- All that's needed:
 - Deterministic actions
 - Fully observable state
 - Initial state, goal state(s)
- What types of problems can be expressed as a graph?



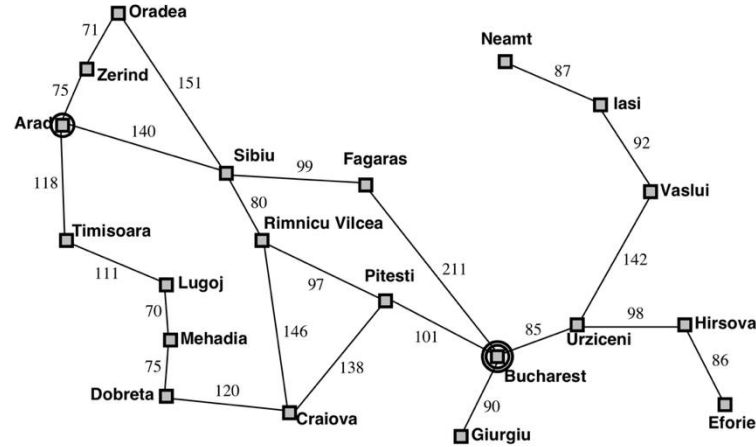
General Form of Search Problems

Household Robot as an Agent



Naïve Formulation of Search Problems

- Given:
 - Graph $\langle V, E, A, \ell \rangle$
 - V set of vertices
 - E set of **directed** edges
 - A set of actions
 - $\ell: E \rightarrow \langle A, \mathbb{R}^+ \rangle$
 - $v_i \in V$ start vertex
 - G goal vertex set
- Find a path from v_i to G



Naïve Formulation of Search Problems

- Given:

- Graph $\langle V, E, A, \ell \rangle$

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- Find a path from v_i

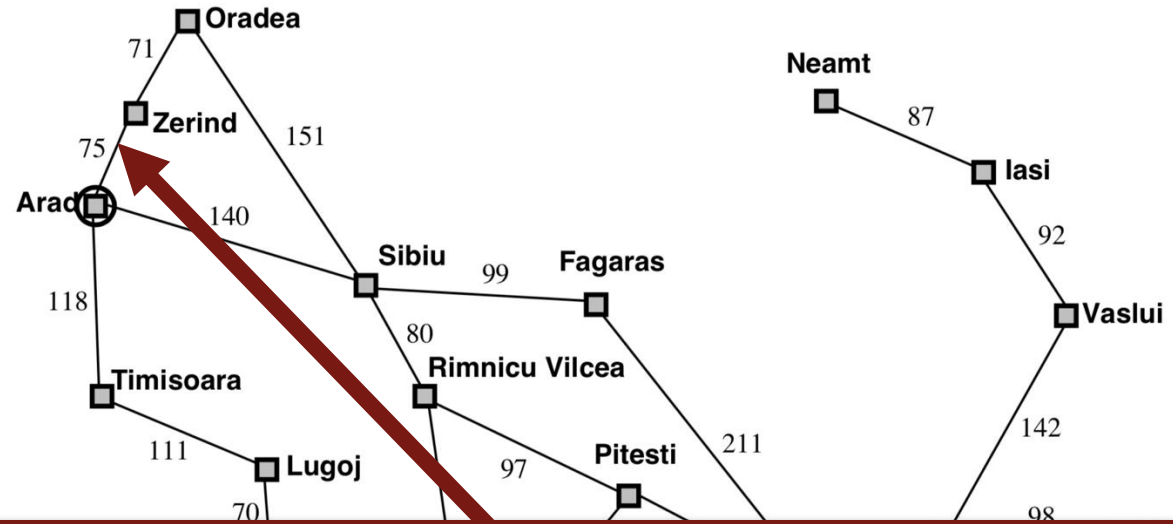
(Arad, Zerind)

(Zerind, Arad)

$\ell((Arad, Zerind)) = \langle drive_Arad_Zerind, 75 \rangle$

- The directed edge for

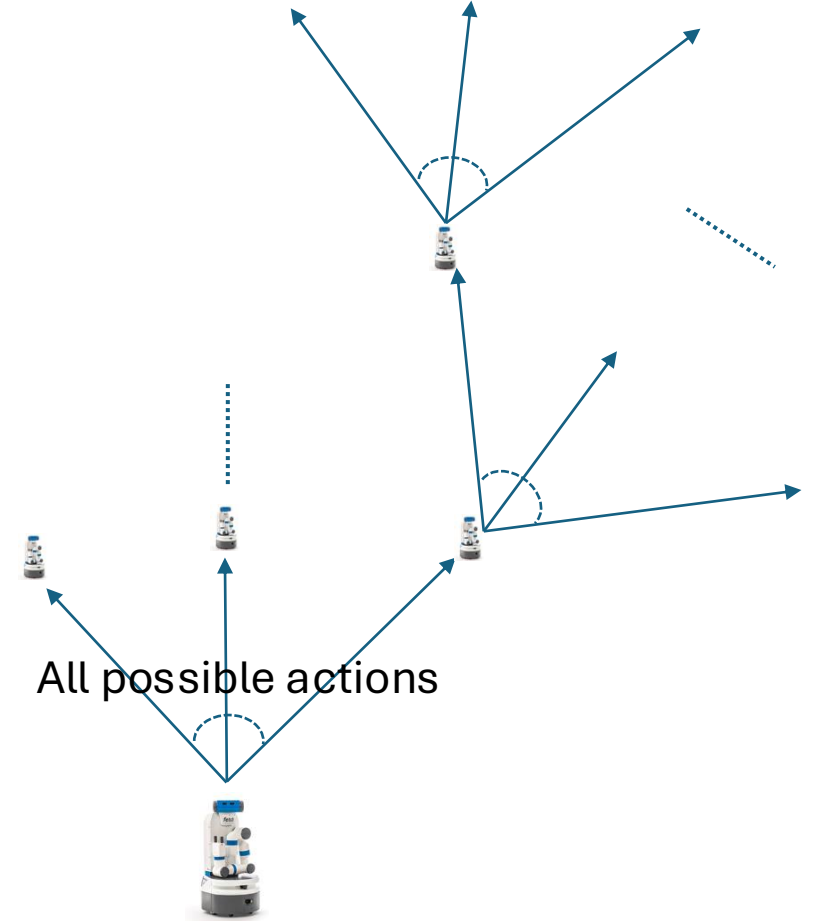
$\ell((Zerind, Arad)) = \langle drive_Zerind_Arad, 75 \rangle$



Problem with Naïve Formulation of Search Problems

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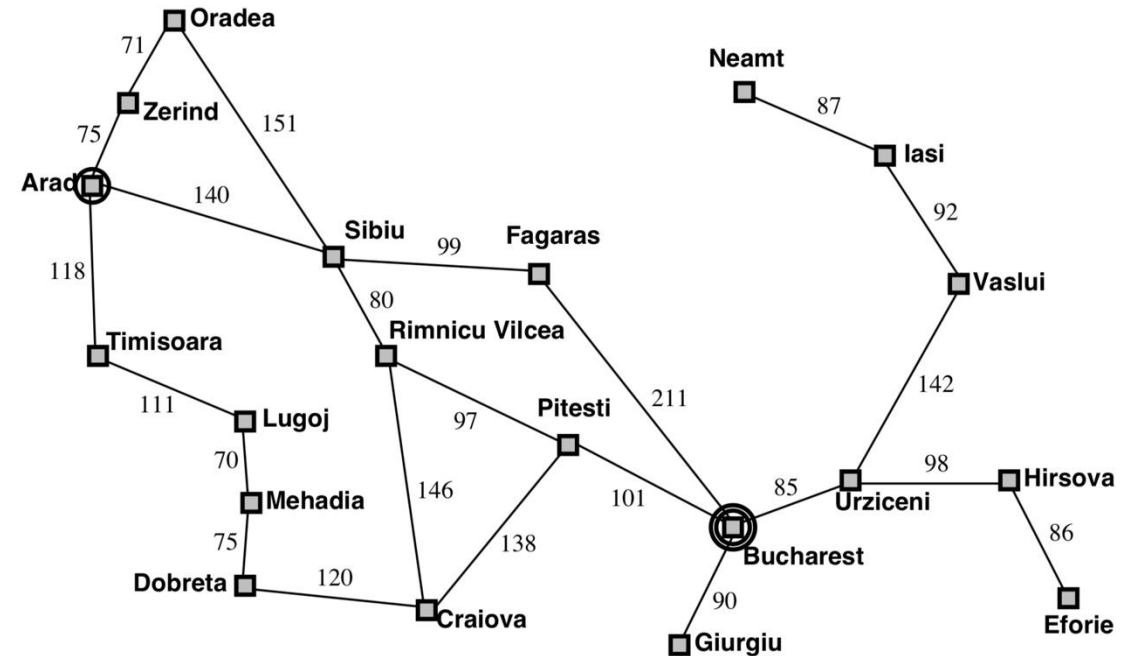
In most cases, the full graph is too large to specify explicitly



Search Problem Defined Over an Implicit Graph

- A **search problem** is defined by
 - S set of possible states
 - Defined, but not enumerated a priori
 - $s_0 \in S$ start state ($\sim$ vertex)
 - A set of actions
 - $N: S \rightarrow S$ successor function
 - $T: S \times A \rightarrow S$ deterministic transition function
 - $N(s) = \{s' : \exists a T(s, a) = s'\}$

Set of all s' that can be reached in one step from s

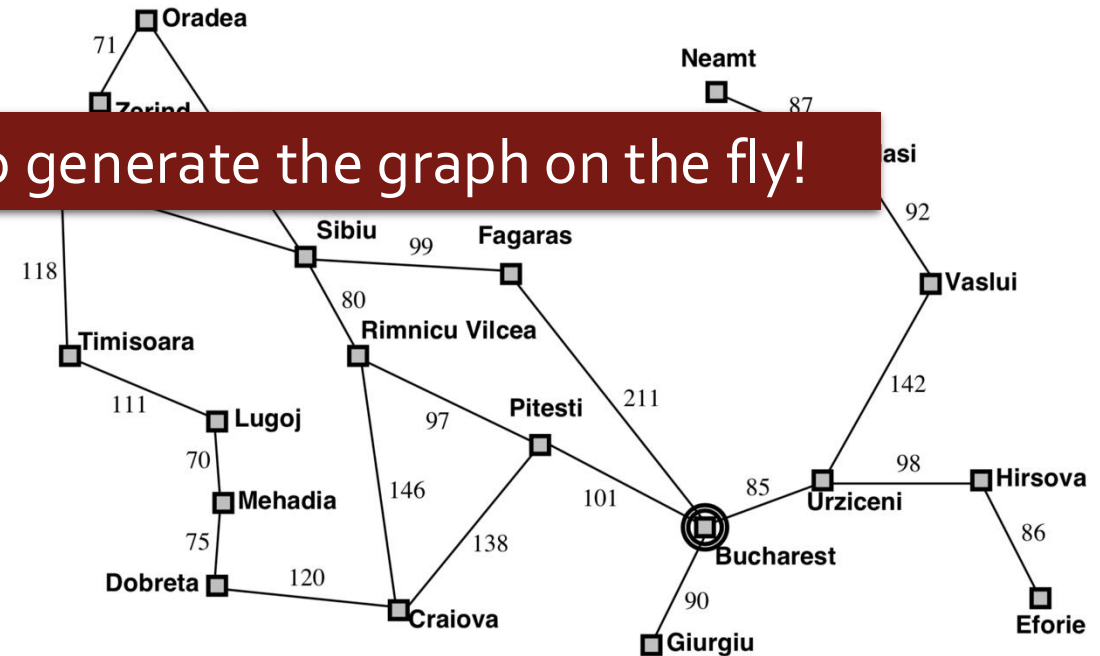


- $N(\text{Arad}) = \{\text{Timisoara}, \text{Sibiu}, \text{Zerind}\}$

Search Problem Defined Over an Implicit Graph

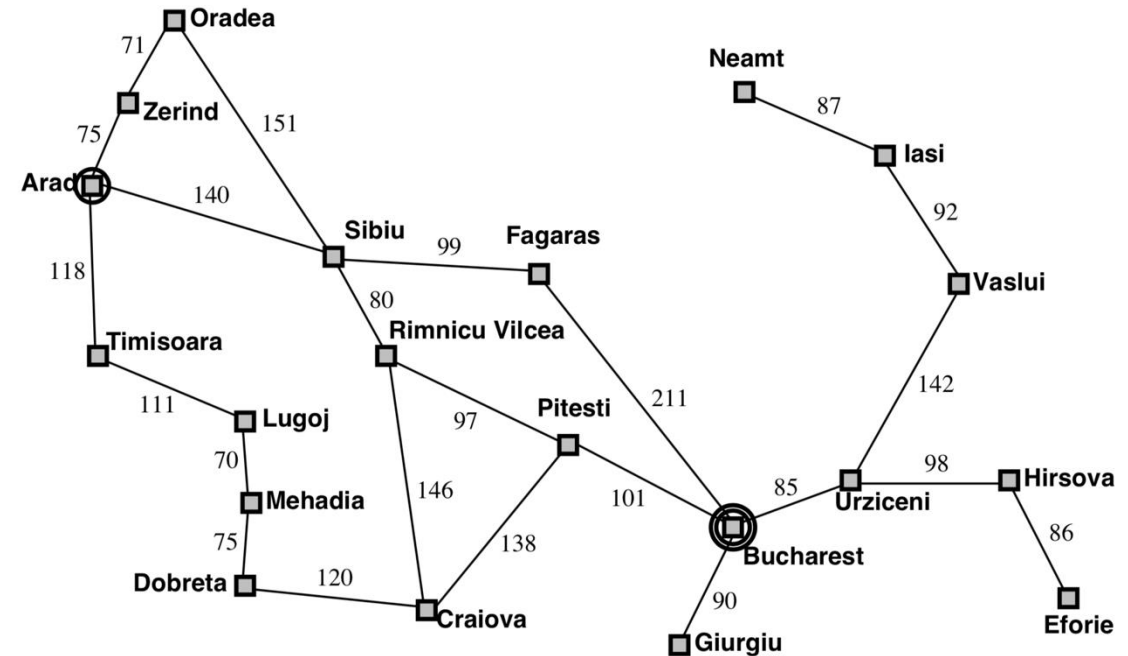
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Allows us to generate the graph on the fly!



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 - $N(s) = \{s' : \exists a T(s, a) = s'\}$
 - $G: S \rightarrow \{True, False\}$ goal test
 - $C: S \times A \times S \rightarrow \mathbb{R}^+$ path cost

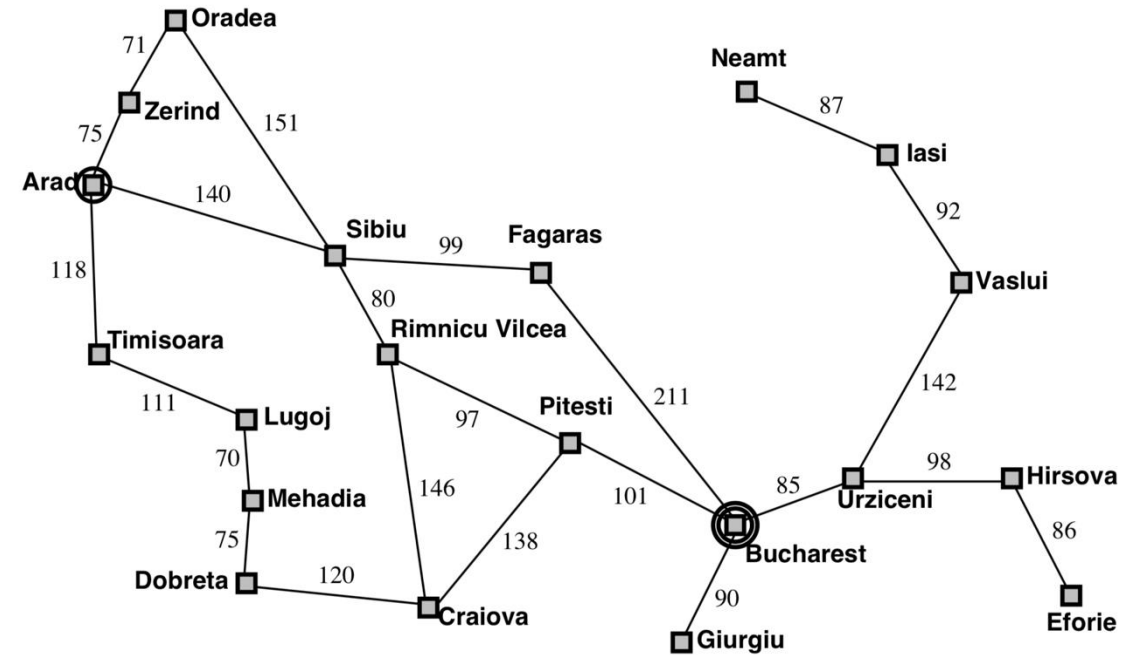


A **solution** to a search problem is a sequence of actions leading from s_0 to a state satisfying the goal test

Role of Abstraction in Modeling Search Problems

- Search problem formulations are models
 - Most models are abstractions
 - Driving from one city to another involves
 - Traffic, one-way streets, fuel considerations, rest stops, ...
- Abstract state = set of real states
- Abstract action = set of procedures

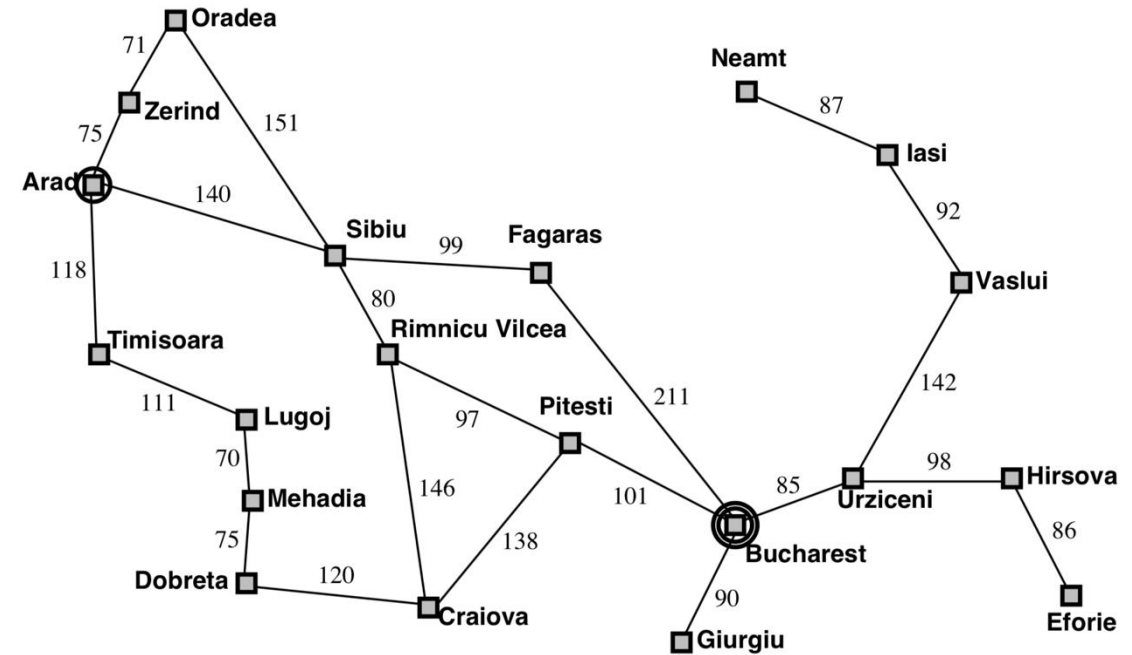
Abstraction is a common and important aspect of modeling



Abstract problem should be easier than the real problem!

Role of Abstraction in Modeling Search Problems

- For the abstract solution to be usable we need:
 - $\forall s, a, s'$ such that $T(s, a) = s'$
 - Every real state in s gets to some real state in s' by following a procedure defined by a
 - “Downward refinability”
 - There should be a way to go from all real locations “in Arad” to some real location “in Zerind”



Abstract problem should be easier than the real problem!